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Discretization in semi-infinite programming:
The rate of approximation

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Abstract

The discretization approach for solving semi-infinite optimization problems is considered. We are interested in the rate of the approximation error between the solution of the semi-infinite problem and the solution of the discretized program depending on the discretization mesh-size d . It will be shown how this rate depends on whether the minimizer is strict of order one or two and on whether the discretization includes boundary points of the index set in a consistent way. This is done for common and for generalized semi-infinite problems.

Keywords: Semi-infinite programming, discretization methods, rate of approximation.

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1 Introduction

A *semi-infinite optimization problem* is a program with infinitely many constraints,

$$\text{SIP: } \min f(x) \quad \text{subject to } x \in F = \{x \in \mathbb{R}^n \mid g(x, y) \leq 0, y \in Y\}$$

where f, g are given real-valued functions and $Y \subset \mathbb{R}^m$ is a compact infinite index set. A generalization to vector-valued constraints g is not difficult.

A common approach for solving SIP are *discretization methods* (see e.g. [8], [4] and the references there). In such a method one chooses a finite discretization Y_d of Y , $Y_d \subset Y$, and solves the finite program

$$\text{SIP}(Y_d): \min f(x) \quad \text{subject to } x \in F(Y_d) = \{x \in \mathbb{R}^n \mid g(x, y) \leq 0, y \in Y_d\} .$$

Throughout the paper d denotes the Hausdorff distance between Y and Y_d ,

$$d := \text{dist}(Y_d, Y) , \quad \text{where } \text{dist}(Y_d, Y) := \max_{y \in Y} \min_{\hat{y} \in Y_d} \|\hat{y} - y\| ,$$

with $\|\cdot\|$, the Euclidean norm. The distance d is a measure for the mesh-size of the discretization.

Now, let \bar{x} be a local solution of SIP and suppose x_d are local solutions of $\text{SIP}(Y_d)$ such that $x_d \rightarrow \bar{x}$ for $d \rightarrow 0$. In this paper we are interested in the rate of approximation, i.e. in the (largest) exponent $\tau > 0$ such that

$$\|x_d - \bar{x}\| \leq O(d^\tau) \quad \text{for } d \rightarrow 0 .$$

This question has been studied in the context of linear Chebyshev approximation (see e.g. [1]). Here, a function f defined on a compact set $Y \subset \mathbb{R}^m$ has to be approximated by a linear combination $p_n(x, y) = \sum_{j=1}^n x_j g_j(y)$ of functions g_j , i.e. we have to solve the problem

$$A(Y): \min_{x \in \mathbb{R}^n} e(x) \quad \text{with } e(x) := \max_{y \in Y} |f(y) - p_n(x, y)| . \quad (1)$$

Note that this problem is a special instance of a semi-infinite problem. In fact, by introducing an extra real variable e the problem $A(Y)$ is equivalent with the linear SIP,

$$\text{SIP}_A: \min_{x, e} e \quad \text{s.t. } \pm (f(y) - p_n(x, y)) \leq e \quad \text{for all } y \in Y .$$

Let again \bar{x} be the solution of $A(Y)$ and x_d the solutions of $A(Y_d)$ on discretizations Y_d of Y . Then the following has been shown in Cheney [1, Chap.3, Sec.6]: If the functions f, g_j are Lipschitz continuous on Y and if the space $\text{span}\{g_1, \dots, g_n\}$ satisfies the so-called Haar condition, then with some $c > 0$,

$$\|x_d - \bar{x}\| \leq c d \quad \text{for } d \text{ small enough} . \quad (2)$$

For the linear Chebyshev approximation on an interval $Y = [-1, 1]$, Dunham observed in [2] that the better estimate

$$||x_d - \bar{x}|| \leq O(d^2)$$

holds if the boundary points ± 1 are contained in the discretization Y_d . A similar result is proven for the complex approximation on a Jordan curve (cf. [3]).

In the present paper we will generalize these results from linear approximation theory to general (nonlinear) semi-infinite problems and improve the estimates.

It is well-known in approximation theory that the Haar condition implies that the solution \bar{x} of $A(Y)$ is strongly unique (strict of order one), i.e. we have with some $q > 0$ (cf. (1)),

$$e(x) - e(\bar{x}) \geq q ||x - \bar{x}|| \quad \text{for all } x \in \mathbb{R}^n .$$

More generally for SIP, a feasible point $\bar{x} \in F$ is said to be a strict local minimizer of order $p = 1$ or $p = 2$ if there is a neighborhood U of \bar{x} and a constant $q > 0$ such that

$$f(x) - f(\bar{x}) \geq q ||x - \bar{x}||^p \quad \text{for all } x \in F \cap U . \quad (3)$$

In the following we will show that the rate of approximation depend firstly on the order of the minimizer and secondly on the structural quality (not merely on the density d) of the discretization. Under weak conditions on SIP, without any assumptions on Y , the following rate holds:

$$||x_d - \bar{x}|| = O(d^{1/p}) \quad \text{for } d \rightarrow 0 .$$

Under additional regularity conditions on Y and if we choose the discretization Y_d such that also points on the boundary of Y are included in Y_d in a consistent way (to be defined later on) then the rate will be proven to behave as

$$||x_d - \bar{x}|| = O(d^{2/p}) \quad \text{for } d \rightarrow 0 .$$

Such inequalities have been obtained in [13] for a vector-valued approximation problem which has been transformed to a semi-infinite problem with index sets $Y = S^m$, the unit sphere in \mathbb{R}^m . A related result has been given earlier by Opfer in [10] where a complex approximation problem led to a SIP with $Y = S^2$.

Remark : In approximation theory and semi-infinite optimization, for certain restricted classes of problems, the solution can be shown to be strictly unique of order one.

For most problem classes however minimizers are of order two. In these cases, in discretization methods, it is important to choose the discretizations Y_d in such a way that the rate

$$||x_d - \bar{x}|| = O(d) \quad \text{for } d \rightarrow 0$$

is obtained and not the slow rate $\|x_d - \bar{x}\| = O(\sqrt{d})$. The latter would lead to an extremely inefficient method.

Note that it can be shown, that generically in SIP, if all problem functions are smooth enough, a minimizer is either strict of order one or strict of order two.

The paper is organized as follows. In Section 2 we briefly survey results in discretization methods related to our study and present a motivating example. In Section 3 we give the main results on the rate of the error $\|x_d - \bar{x}\|$ for problems SIP. In Section 4 the analysis is extended to semi-infinite problems with variable index sets, also called generalized SIP.

2 Discretization method and an example

In this section we briefly describe the discretization method for solving SIP and give some results related to the investigations of the paper.

A conceptual discretization method proceeds as follows.

Discretization method : Choose a sequence of finite discretizations $Y_k \subset Y$ such that $Y_0 \subset Y_k$ and $\text{dist}(Y_k, Y) \rightarrow 0$, for $k \rightarrow \infty$. Choose a (small) $\varepsilon > 0$ and put $k = 0$.

Step k: Compute a solution x_k of $\text{SIP}(Y_k)$. If x_k is nearly feasible for SIP in the sense that $g(x_k, y) \leq \varepsilon$, $y \in Y$, we stop with the approximate solution $\hat{x} = x_k$.

Otherwise, go to step $k + 1$.

For a discussion of this method the reader is referred to [8], [11] and for the related exchange methods to [5], [8]. Implementations of such discretization methods are developed in [6] and [7].

Note that the problem $\text{SIP}(Y_*)$ has a (global) solution if the corresponding feasible set $F(Y_*)$ is bounded (compact). So, for the discretization method (and appropriate exchange methods) the following convergence result holds (see e.g. [8] for a proof).

Theorem 1 *Suppose the feasible set $F(Y_0)$ corresponding to the coarsest grid Y_0 is compact. Then all problems $\text{SIP}(Y_k)$ have solutions $x_k \in F(Y_0)$ and the sequence $\{x_k\}$ has an accumulation point \bar{x} . Each such accumulation point is a solution of SIP.*

In the present paper we do not investigate under which conditions solutions x_k of $\text{SIP}(Y_k)$ converge to a solution \bar{x} of SIP. In this direction the following has been proven in [5] for linear SIP, i.e. $f(x) = c^T x$ and $g(x, y) = a^T(y)x - b(y)$ (a, b continuous):
If

$$-c \in \text{int}(\text{cone}\{a(y), y \in Y\})$$

then to each solution x_k of $\text{SIP}(Y_k)$ there is a solution \bar{x}_k of SIP such that

$$\|x_k - \bar{x}_k\| \rightarrow 0 \quad \text{for } k \rightarrow \infty .$$

In this paper the following assumption will always hold.

A1 : We have given a problem SIP and discretizations Y_d of Y such that with $d := \text{dist}(Y, Y_d)$ the limit $d \rightarrow 0$ holds. The point \bar{x} is a local minimizer of SIP and x_d are local solutions of $\text{SIP}(Y_d)$ such that

$$\|x_d - \bar{x}\| \rightarrow 0 \quad \text{for } d \rightarrow 0 .$$

To illustrate the main idea behind the general analysis in the next section we work out an example. To obtain the improved rate

$$\|x_d - \bar{x}\| = O(d^{2/p}) \quad \text{for } d \text{ small} ,$$

the boundary of Y will play an important role.

For later purposes we define for feasible \bar{x} the active index set

$$Y_0(\bar{x}) = \{\bar{y} \in Y \mid g(\bar{x}, \bar{y}) = 0\} .$$

Let us consider the semi-infinite program

$$\text{SIP : } \min f(x) := -x_1 + x_2 \quad \text{s.t.} \quad x_1(1-y)(1+y) + y^2(y^2 - x_2) \leq 0, \quad y \in Y ,$$

where $Y = [-1, 1]$. We directly find $x_1 \leq 0$ and $x_2 \geq 1$ and for any feasible x with $\bar{x} := (0, 1)$,

$$f(x) - f(\bar{x}) = -x_1 + x_2 - 1 = |x_1| + |x_2 - 1| \geq \|x - \bar{x}\| .$$

Consequently, the point $\bar{x} = (0, 1)$ is the strict minimizer of order one. The active index set is

$$Y_0(\bar{x}) = \{0, -1, 1\} .$$

Consider now the semi-infinite problem $\text{SIP}(Y_d)$ on the discretization $Y_d = \{\pm d, \pm 2d, \dots, \pm(n-1)d\}$, with $d := \frac{1}{n}$. The solution x_d is given by the relation (fill in $y = \pm d, y = \pm(1-d)$)

$$\begin{aligned} x_1(1-d^2) + d^4 - d^2 x_2 &= 0 \\ x_1 d(2-d) + (1-d)^4 - (1-d)^2 x_2 &= 0 . \end{aligned}$$

We find $x_d = (x_1, x_2)$ with

$$x_2 = \frac{(1-d)^4 - \frac{d^5(2-d)}{(1-d)^2}}{(1-d)^2 - \frac{d^3(2-d)}{1-d^2}} = 1 - 2d + o(d), \quad x_1 = \frac{d^2}{1-d^2} x_2 - \frac{d^4}{(1-d)^2} = d^2 + o(d^2)$$

and the linear rate

$$\|x_d - \bar{x}\| = 2d + o(d) .$$

Now, we add the boundary points ± 1 of Y to the discretization, $\hat{Y}_d = Y_d \cup \{-1, 1\}$. Recall that $y = \pm 1$ are index points, active at the minimizer \bar{x} . The solution of $\text{SIP}(\hat{Y}_d)$ is defined by the relation (fill in $y = \pm d, y = \pm 1$)

$$\begin{aligned} x_1(1-d^2) + d^4 - d^2 x_2 &= 0 \\ 1 - x_2 &= 0 . \end{aligned}$$

Now, we obtain the solution $\hat{x}_d = (\hat{x}_1, \hat{x}_2)$ with $\hat{x}_2 = 1$, $\hat{x}_1 = \frac{d^2-d^4}{1-d^2} = d^2 + o(d^2)$ and the quadratic rate

$$\|\hat{x}_d - \bar{x}\| = d^2 + o(d^2) .$$

This example shows that the active index points on the boundary of Y play a more crucial role than the inner active points.

To understand this phenomenon, let us fix a solution x_d of $\text{SIP}(Y_d)$ (which is not feasible for SIP). Consider the problem

$$\max_y g(x_d, y) \quad \text{s.t.} \quad y \in Y = [-1, 1] . \quad (4)$$

Suppose the solution y_d lies in the interior of Y . Then

$$D_y g(x_d, y_d) = 0 . \quad (5)$$

Choose now a discretization point $\hat{y}_d \in Y_d$ such that $|\hat{y}_d - y_d| \leq d$, which is possible by the definition of d . Then, since $g(x_d, \hat{y}_d) \leq 0$ we find (for $g \in C^2$)

$$\begin{aligned} g(x_d, y_d) &\leq g(x_d, y_d) - g(x_d, \hat{y}_d) \\ &= D_y g(x_d, y_d)(y_d - \hat{y}_d) + \frac{1}{2} D_y^2 g(x_d, y_d)(y_d - \hat{y}_d)^2 + o((y_d - \hat{y}_d)^2) \\ &= O((y_d - \hat{y}_d)^2) = O(d^2) . \end{aligned} \quad (6)$$

If the solution y_d of (4) is a boundary point $y = \pm 1$ and y_d is not included in Y_d (i.e. possibly $|\hat{y}_d - y_d| = d$) then instead of (5) (in general) we have $D_y g(x_d, y_d) \neq 0$ and instead of (6) we only obtain

$$g(x_d, y_d) \leq g(x_d, y_d) - g(x_d, \hat{y}_d) = D_y g(x_d, y_d)(y_d - \hat{y}_d) + o((y_d - \hat{y}_d)) = O(d) .$$

So, it is crucial to add ‘boundary points’ to the discretization of Y .

In the next section, also for index sets Y in higher dimensions, it is shown how the approximation rate $\|x_d - \bar{x}\|$ can be doubled if discretization points on the boundary of Y are included into the discretization Y_d in a certain way.

3 Rate of approximation

Let throughout this section the assumption A1 hold. We are now going to analyze how fast the error $\|x_d - \bar{x}\|$ tends to zero with the mesh-size d .

The function $g(x, y)$ is said to be Lipschitz continuous near \bar{x} w.r.t. Y if there exist $\varepsilon > 0$ and $L > 0$ such that

$$|g(x, y) - g(x, \hat{y})| \leq L \|y - \hat{y}\| \quad \text{for all } y, \hat{y} \in Y, \quad \|x - \bar{x}\| < \varepsilon .$$

Note that this condition is fulfilled if g is continuously differentiable on $\mathbb{R}^n \times \mathbb{R}^m$. We begin with a general lemma.

Lemma 1 *Let g be Lipschitz continuous near \bar{x} w.r.t. Y . Then there exist $\delta > 0$, $L > 0$ such that for all d , $0 < d < \delta$ we have*

$$g(x_d, y) \leq L d \quad \text{for all } y \in Y .$$

PROOF : Let y_d be a solution of $\max_{y \in Y} g(x_d, y)$ and let \hat{y}_d be a point in Y_d such that $\|\hat{y}_d - y_d\| \leq d$. By Lipschitz continuity of g and using $g(x_d, \hat{y}_d) \leq 0$ we find

$$g(x_d, y_d) \leq g(x_d, y_d) - g(x_d, \hat{y}_d) \leq L \|\hat{y}_d - y_d\| \leq L d$$

with $L > 0$ for $\|x_d - \bar{x}\|$ small enough. \square

Under additional regularity conditions on Y and if, roughly speaking, the boundary of Y is also taken in consideration, the inequality in Lemma 1 can be sharpened to

$$g(x_d, y) \leq L d^2 \quad \text{for all } y \in Y ,$$

(cf. the illustrative example of Section 2). To do so we have to introduce assumptions on g, Y, Y_d .

A2 : Let the following hold:

- (α) There is a neighborhood \bar{U} of \bar{x} such that the function $D_y^2 g(x, y)$ is continuous on $\bar{U} \times Y$. (This implies that g is Lipschitz continuous near \bar{x} w.r.t. Y .)
- (β) The index set $Y \subset \mathbb{R}^m$ is compact, non-empty and explicitly given as the solution set of inequalities,

$$Y = \{y \in \mathbb{R}^m \mid v_i(y) \leq 0, i \in I\} ,$$

where I is a finite index set and $v_i \in C^2(Y)$.

- (γ) The *Mangasarian Fromovitz Constraint Qualification* (MFCQ) holds for Y : For any $\bar{y} \in Y$ with the *active index set* $I(\bar{y}) := \{i \in I \mid v_i(\bar{y}) = 0\}$ there exists a vector $\bar{\eta} = \eta(\bar{y})$ such that

$$Dv_i(\bar{y})\bar{\eta} < 0, \quad i \in I(\bar{y}) . \tag{7}$$

The *Linear Independency Constraint Qualification* (LICQ) is said to hold for Y , if for any $\bar{y} \in Y$ the vectors

$$Dv_i(\bar{y}), \quad i \in I(\bar{y}) , \quad \text{are linearly independent} . \tag{8}$$

Note that the condition LICQ is stronger than MFCQ.

Consider for fixed x under assumptions A2 the so-called lower level problem

$$Q(x) : \quad \max_y g(x, y) \quad \text{s.t. } v_i(y) \leq 0, i \in I .$$

By MFCQ (see (7)), at a solution \bar{y} of $Q(x)$ necessarily the Kuhn-Tucker condition

$$D_y g(x, \bar{y}) - \sum_{i \in I(\bar{y})} \mu_i Dv_i(\bar{y}) = 0$$

holds with Lagrange multipliers $\mu_i \geq 0$. We need a result on the uniform boundedness of the Lagrange multipliers.

Lemma 2 Suppose, the assumption A2 holds for Y, g, v_i . Let $C \subset \mathbb{R}^n$ be compact. Then, there exists $\alpha > 0$ such that for any $x \in C, y \in Y$ and $\mu \geq 0$ satisfying

$$D_y g(x, y) - \sum_{i \in I(y)} \mu_i Dv_i(y) = 0$$

we have $\|\mu\| \leq \alpha$.

PROOF : Assume that on the contrary there are sequences $\{x_k\}, \{y_k\}, \{\mu^k\}$, $x_k \in C, y_k \in Y$ and $\mu^k \geq 0$ such that

$$D_y g(x_k, y_k) - \sum_{i \in I(y_k)} \mu_i^k Dv_i(y_k) = 0 \quad (9)$$

and $\|\mu^k\| \rightarrow \infty$ for $k \rightarrow \infty$. We divide (9) by $\|\mu^k\|$ and choose an appropriate subsequence such that $x_{k_v} \rightarrow \hat{x} \in C, y_{k_v} \rightarrow \hat{y} \in Y, I_0 = I(y_{k_v})$ (same index set) and $\mu_i^{k_v} / \|\mu^{k_v}\| \rightarrow \hat{\mu}_i, i \in I_0$. By continuity it follows $I_0 \subset I(\hat{y})$ and we find from (9) for $k_v \rightarrow \infty$,

$$-\sum_{i \in I_0} \hat{\mu}_i Dv_i(\hat{y}) = 0$$

with $\hat{\mu} \neq 0$. By multiplying this relation with the MFCQ-vector $\hat{\eta}$ in A2(γ), using $\hat{\mu} \geq 0$, we are led to the contradiction

$$0 = -\sum_{i \in I_0} \hat{\mu}_i Dv_i(\hat{y}) \hat{\eta} > 0.$$

□

Motivated by the example in Section 2 we now pay attention to the boundary of the index set Y and consider all boundary parts of all 'dimensions' in Y .

To that end let us fix a subset $I_* \subset I$ such that there exists a point $y_* \in Y$ with $I_* = I(y_*)$. The set

$$f_* := \{y \in Y \mid v_i(y) = 0, i \in I_*\}$$

defines a subset of Y . There are finitely many such index sets I_* , say I_1, I_2, \dots, I_k with corresponding subsets of Y ,

$$f_l := \{y \in Y \mid v_i(y) = 0, i \in I_l\}, \quad l = 1, \dots, k, \quad (10)$$

satisfying $f_l \cap Y \neq \emptyset$. For $I_1 := \emptyset$, for example, f_1 defines the whole set Y . If $I_l \neq \emptyset$, then f_l defines a boundary-part of Y . Note that since $Y \neq \emptyset$, for $\bar{y} \in Y$, by MFCQ the system $Dv_i(\bar{y})\eta \leq -1, i \in I(\bar{y})$, has a solution η . Thus, in view of

$$v_i(\bar{y} + t\eta) = v_i(\bar{y}) + tDv_i(\bar{y})\eta + o(t) \leq -t + o(t) < 0, \quad i \in I(\bar{y}),$$

for all sufficiently small $t > 0$, (every component of) the set Y has interior points.

Remark : Suppose that instead of the assumption MFCQ in A2(γ) the stronger LICQ holds for Y . Then, consider any point $\bar{y} \in f_l$, $l \in \{1, \dots, k\}$ with $I_l = I(\bar{y})$. By LICQ, the gradients $Dv_i(\bar{y})$, $i \in I_l$, are linearly independent and by using the implicit function theorem, locally near \bar{y} , the set f_l defines a manifold in \mathbb{R}^m of dimension $m - |I_l|$. With regard to the geometry of polyhedra we can call these sets f_l faces of Y of dimension $m - |I_l|$.

Under the weaker MFCQ the sets f_l may have a more complicated structure. We however will use the term 'faces' for the sets f_l also in this case.

We introduce an assumption on the discretization Y_d .

A3 : Suppose Y satisfies A2 and the sets f_l , $l = 1, \dots, k$, (see (10)) define the subsets of Y as constructed above. We assume that the discretization Y_d is chosen such that

$$\max_{y \in f_l} \min_{\hat{y} \in Y_d \cap f_l} \|\hat{y} - y\| \leq d \quad \text{for all } l = 1, \dots, k, \text{ and all } d.$$

This means that on all 'faces' of Y of all 'dimensions', discretization points are chosen with mesh-size d .

Rmark : For the index set $Y = \{y \in \mathbb{R}^2 \mid |y_i| \leq 1, i = 1, 2\}$ for instance, a discretization Y_d in A3 means, that points on the faces ' $y_i = \pm 1$ ' have to be included and the 4 vertices $y = (\pm 1, \pm 1)$ must be added to Y_d .

We are now able to sharpen the result of Lemma 1.

Theorem 2 *Let A1, A2 and A3 hold. Then with the local solutions x_d of SIP(Y_d) for some $L > 0$ we have for small d ,*

$$\max_{y \in Y} g(x_d, y) \leq L d^2.$$

PROOF : Given d small, let y_d be a solution of

$$\max_y g(x_d, y) \quad \text{s.t.} \quad v_i(y) \leq 0, i \in I.$$

Then y_d is contained in some face f_l , $l = l(d)$, with $I_l = I(y_d)$ and y_d must also be a local solution of

$$\max_y g(x_d, y) \quad \text{s.t.} \quad v_i(y) \leq 0, i \in I_l.$$

By MFCQ (see A2(γ)), y_d satisfies the Kuhn-Tucker condition

$$D_y g(x_d, y_d) - \sum_{i \in I_l} \mu_i^d Dv_i(y_d) = 0$$

with $\mu_d \geq 0$. By Lemma 2 the sequence μ^d is bounded, $\|\mu^d\| \leq K$ for $d \rightarrow 0$. In view of A3 (for any d) there is a discretization point $\hat{y}_d \in f_l \cap Y_d$ such that $\|\hat{y}_d - y_d\| \leq d$. Thus using $v_i(y_d) = v_i(\hat{y}_d) = 0$, $i \in I_l$, for d small we find

$$\begin{aligned}
g(x_d, y_d) &\leq g(x_d, y_d) - g(x_d, \hat{y}_d) - \sum_{i \in I_l} \mu_i^d \left(v_i(y_d) - v_i(\hat{y}_d) \right) \\
&= \left(D_y g(x_d, y_d) - \sum_{i \in I_l} \mu_i^d D v_i(y_d) \right) (y_d - \hat{y}_d) \\
&\quad + \frac{1}{2} (y_d - \hat{y}_d)^T \left(D_y^2 g(x_d, y_d) - \sum_{i \in I_l} \mu_i^d D^2 v_i(y_d) \right) (y_d - \hat{y}_d) \\
&\quad + o(\|y_d - \hat{y}_d\|^2) \\
&\leq L \|y_d - \hat{y}_d\|^2 \leq L d^2
\end{aligned}$$

with some $L > 0$. □

In the next step, from the solution x_d of $\text{SIP}(Y_d)$ we construct a nearby point \hat{x}_d which is feasible for $\text{SIP}(Y)$. To achieve this we assume that the *Mangasarian Fromovitz Constraint Qualification* (MFCQ) is valid for \bar{x} : There there exists a vector $\xi \in \mathbb{R}^m$ such that

$$D_x g(\bar{x}, \bar{y}) \xi \leq -\kappa < 0, \quad \text{for all } \bar{y} \in Y_0(\bar{x}), \quad (11)$$

with $Y_0(\bar{x})$, the active index set. Note that for semi-infinite formulations of Chebyshev approximation problems the MFCQ is automatically fulfilled at any feasible point \bar{x} .

A4 : We assume that $D_x g(x, y)$ is continuous on $\bar{U} \times Y$, where \bar{U} is a neighborhood of \bar{x} . Let moreover MFCQ be valid at the local minimizer \bar{x} of SIP.

Lemma 3 Suppose, A4 holds and with some $\gamma > 0$, $c_1 > 0$ we have

$$\max_{y \in Y} g(x_d, y) \leq c_1 d^\gamma, \quad \text{for } d \text{ small}.$$

Then there exists $\rho > 0$ such that with the MFCQ-vector ξ in (11), for small d , the points $\hat{x}_d := x_d + \rho d^\gamma \xi$ are feasible for SIP, i.e.

$$g(\hat{x}_d, y) \leq 0, \quad y \in Y.$$

PROOF : We proceed in two steps. Firstly, for $\varepsilon > 0$ we consider the relative open set $Y_0^\varepsilon(\bar{x}) := \{y \in Y \mid \|y - \bar{y}\| < \varepsilon \text{ for some } \bar{y} \in Y_0(\bar{x})\}$. By MFCQ using the continuity of $D_x g$ there is some $\varepsilon > 0$ such that

$$D_x g(x, y) \xi \leq -\frac{\kappa}{2} \quad \text{for } y \in Y_0^\varepsilon(\bar{x}), \quad \|x - \bar{x}\| < \varepsilon.$$

Thus if $||\rho d^\gamma \xi|| < \varepsilon$, for all $y \in Y_0^\varepsilon(\bar{x})$ and small d we find,

$$\begin{aligned} g(\hat{x}_d, y) &= g(x_d, y) + \rho d^\gamma D_x g(x_d, y) \xi + o(\rho d^\gamma) \\ &\leq c_1 d^\gamma - \rho \frac{\kappa}{2} d^\gamma + o(\rho d^\gamma) \\ &= d^\gamma (c_1 - \rho \frac{\kappa}{2}) + o(\rho d^\gamma) \leq 0 \end{aligned} \tag{12}$$

if we choose $\rho, \rho \frac{\kappa}{2} > c_1$.

Secondly, we consider the compact set $Y \setminus Y_0^\varepsilon(\bar{x})$. By continuity of g for given $\varepsilon > 0$ there exists $\varepsilon_1 > 0$ such that

$$g(x, y) < 0 \quad \text{for all } y \in Y \setminus Y_0^\varepsilon(\bar{x}), \quad ||x - \bar{x}|| < \varepsilon_1 .$$

Together with (12) we find $g(\hat{x}_d, y) \leq 0$, $y \in Y$, for \hat{x}_d if

$$||\hat{x}_d - \bar{x}|| = ||x_d - \bar{x} + \rho d^\gamma \xi|| \leq ||x_d - \bar{x}|| + \rho d^\gamma ||\xi|| < \min\{\varepsilon, \varepsilon_1\} .$$

□

Note that since Y_d is contained in Y , for the object values of \bar{x} and x_d we trivially have

$$f(\bar{x}) \geq f(x_d) .$$

We also will assume that the object function f is Lipschitz continuous near \bar{x} : There exist $\varepsilon > 0$ and $L_f > 0$ such that

$$|f(\hat{x}) - f(x)| \leq L_f ||\hat{x} - x|| \quad \text{for all } ||\hat{x} - \bar{x}||, ||x - \bar{x}||, < \varepsilon .$$

By combining the preceding lemmas we obtain a result on the error $f(\bar{x}) - f(x_d)$.

Corollary 1 *Suppose A1, A4 hold and f is Lipschitz continuous near \bar{x} . Then we have.*

(a) *There is some $c_2 > 0$ such that for all d small enough,*

$$0 \leq f(\bar{x}) - f(x_d) \leq c_2 d .$$

(b) *If in addition, A2 and A3 is satisfied for Y, Y_d , then with some $c_3 > 0$ the inequality*

$$0 \leq f(\bar{x}) - f(x_d) \leq c_3 d^2$$

is valid for small d .

PROOF : By MFCQ, using Lemma 1, Lemma 3 we find with $\gamma = 1$ that for some $\rho > 0$ the points $x_d + \rho d^\gamma \xi$ are feasible,

$$g(x_d + \rho d^\gamma \xi, y) \leq 0, \quad y \in Y,$$

if d is small. This implies by Lipschitz continuity of f ,

$$f(\bar{x}) - f(x_d) \leq f(x_d + \rho d^\gamma \xi) - f(x_d) = O(d^\gamma)$$

Under the assumptions A2, A3 by using Theorem 2 instead of Lemma 1 we obtain the same relation with $\gamma = 2$. \square

The behavior of the error $\|x_d - \bar{x}\|$ depends on the type of the local minimizer \bar{x} , namely on the fact whether \bar{x} is a strict local minimizer of order $p = 1$ or of order $p = 2$.

Theorem 3 *Suppose that the assumptions of Corollary 1 hold. Then if \bar{x} is a strict local minimizer of order $p = 1$ or $p = 2$ we have:*

(a) *There is some $\sigma_2 > 0$ such that for all d*

$$\|x_d - \bar{x}\| \leq \sigma_1 d^{1/p}.$$

(b) *If in addition A2 and A3 is satisfied for Y, Y_d , then with some $\sigma_2 > 0$ the inequality*

$$\|x_d - \bar{x}\| \leq \sigma_2 d^{2/p}$$

is valid.

PROOF : By Corollary 1, with $\gamma = 1$ in (a) or $\gamma = 2$ in (b) the vectors $x_d + \rho d^\gamma \xi$ are feasible for SIP and with the factor q in (3) we obtain

$$\begin{aligned} \|x_d + \rho d^\gamma \xi - \bar{x}\|^p &\leq \frac{f(x_d + \rho d^\gamma \xi) - f(\bar{x})}{q} \\ &= \frac{f(x_d + \rho d^\gamma \xi) - f(x_d)}{q} - \frac{f(\bar{x}) - f(x_d)}{q} = O(d^\gamma). \end{aligned}$$

Finally, we find in all cases $p, \gamma = 1, 2$ the relation

$$\|x_d - \bar{x}\| \leq \|x_d + \rho d^\gamma \xi - \bar{x}\| + \|\rho d^\gamma \xi\| = O(d^{\gamma/p}) + O(d^\gamma) = O(d^{\gamma/p}).$$

\square

Remark : In our analysis, for brevity, we did not consider equality constraints for the definition of Y (and/or F). If for example the index set Y is defined by

$$Y = \{y \in \mathbb{R}^m \mid u_j(y) = 0, \quad j \in J, \quad v_i(y) \leq 0, \quad i \in I\},$$

then, all results of this section remain valid if the condition MFCQ in (7) is modified as follows: For any $\bar{y} \in Y$ the vectors $Du_j(\bar{y})$, $j \in J$ are linearly independent and there exists a vector $\bar{\eta} = \eta(\bar{y})$ such that

$$Du_j(\bar{y}) \bar{\eta} = 0, \quad j \in J, \quad Dv_i(\bar{y}) \bar{\eta} < 0, \quad i \in I(\bar{y}).$$

Remark : It is clear from the analysis in this section, that it is sufficient to choose the discretization points only in the neighborhood of the active point set $Y_0(\bar{x})$ (not known in advance). This observation is important if for solving SIP we use an exchange method which in each step implicitly calculates approximations of the active index points.

Remark : As a special case, the following result is contained in the analysis above. Consider for $\varepsilon > 0$ the problem

$$\text{SIP}_\varepsilon: \quad \min f(x) \quad \text{subject to } x \in F_\varepsilon = \{x \in \mathbb{R}^n \mid g(x, y) \leq \varepsilon, \quad y \in Y\}.$$

This means, that the constraints only have to be satisfied approximately. Assume that with a local minimizer \bar{x} of SIP of order $p = 1$ or $p = 2$ and local solutions x_ε of $\text{SIP}(Y_\varepsilon)$ we have $x_\varepsilon \rightarrow \bar{x}$ for $\varepsilon \rightarrow 0$. Then with some $\sigma > 0$ the following rate holds:

$$\|x_\varepsilon - \bar{x}\| \leq \sigma \varepsilon^{1/p}$$

4 Semi-infinite problems with variable index sets

In this section we extend the previous results to the case of semi-infinite problems of the form

$$\text{GSIP:} \quad \min f(x) \quad \text{s.t. } x \in F = \{x \in \mathbb{R}^n \mid g(x, y) \leq 0, \quad y \in Y(x)\}$$

with index sets $Y(x) \subset \mathbb{R}^m$ depending on x . Such programs are also called *generalized semi-infinite problems*.

These semi-infinite problems with variable index sets may have a much more complicated structure than common SIP. For a discussion of GSIP we refer e.g. to [12] and [14]. For example, the feasible set F of GSIP needs not to be closed if the set-valued mapping $Y(x)$ is not continuous. So in the following we will assume:

$$Y : \mathbb{R}^n \rightarrow 2^{\mathbb{R}^m}, \quad Y(x) \subset C_0, \quad \text{for all } x \in \mathbb{R}^n \quad \text{with } C_0 \subset \mathbb{R}^m \text{ compact} \quad (13)$$

and that the mapping Y is continuous on \mathbb{R}^n .

Discretization methods for GSIP have been considered in [15]. Let as for SIP be chosen finite discretizations $Y_d(x) \subset Y(x)$, $x \in \mathbb{R}^n$, and let $\text{GSIP}(Y_d)$ denote the finite problem with the index-set $Y_d(x)$ and corresponding feasible set $F(Y_d)$. To have a closed feasible set $F(Y_d)$ also the discretizations $Y_d(x)$ must be continuous in x (see [15] for details). So, we assume, that the discretizations are given by

$$Y_d(x) = \{y_d^j(x), j = 1, \dots, j_d\}, \quad x \in \mathbb{R}^n,$$

with continuous functions $y_d^j : \mathbb{R}^n \rightarrow C_0$, $y_d^j(x) \in Y(x)$. For a result as in Theorem 1 we refer to [15, Th.2].

As before we assume

A1_G : Let \bar{x} be a local minimizer of GSIP and let x_d be local solutions of $\text{GSIP}(Y_d)$ such that

$$\|x_d - \bar{x}\| \rightarrow 0 \quad \text{for } d \rightarrow 0.$$

Here we assume that $N_{\bar{x}} \subset \mathbb{R}^n$ is a (fixed) compact set with \bar{x} as inner point and d is defined by

$$d = \max_{x \in N_{\bar{x}}} \text{dist}(Y_d(x), Y(x)).$$

We are again interested in the rate of the error $\|x_d - \bar{x}\|$ and we are going to generalize the results of Section 3 to semi-infinite problems with variable index sets. Firstly we extend Lemma 1.

Lemma 4 *Let g be Lipschitz continuous near \bar{x} w.r.t. C_0 . Then there exist $\delta > 0$, $L > 0$ such that for all d , $0 < d < \delta$ we have*

$$g(x_d, y) \leq L d \quad \text{for all } y \in Y(x_d).$$

PROOF : Let y_d be a solution of $\max_{y \in Y(x_d)} g(x_d, y)$ and let \hat{y}_d be a point in $Y_d(x_d)$ such that $\|\hat{y}_d - y_d\| \leq d$. By Lipschitz continuity of g we find $g(x_d, y_d) \leq L d$ as in the proof of Lemma 1. \square

The assumption A2 has to be modified.

A2_G : Let the following hold:

(α) The condition as in A2(α) is satisfied with Y replaced by C_0 .

(β) The index set $Y(x)$ is given by

$$Y(x) = \{y \in \mathbb{R}^m \mid v_i(x, y) \leq 0, i \in I\},$$

where I is a finite index set and $v_i \in C^2(\mathbb{R}^n \times C_0)$.

(γ) MFCQ holds for $Y(\bar{x})$: For any $\bar{y} \in Y(\bar{x})$ there exists a vector $\bar{\eta} = \eta(\bar{y})$ such that

$$D_y v_i(\bar{x}, \bar{y}) \bar{\eta} < 0, \quad i \in I(\bar{x}, \bar{y}) := \{i \in I \mid v_i(\bar{x}, \bar{y}) = 0\}. \quad (14)$$

LICQ is modified accordingly.

It is not difficult to show that MFCQ (and LICQ) are stable in the following sense (we omit the proof).

Lemma 5 *Let MFCQ hold for $Y(\bar{x})$. Then there exists $\varepsilon_0 > 0$ such that for any \hat{x} , $\|\hat{x} - \bar{x}\| \leq \varepsilon_0$ and any $\hat{y} \in Y(\hat{x})$ there is a vector $\hat{\eta}$ satisfying*

$$D_y v_i(\hat{x}, \hat{y}) \hat{\eta} < 0, \quad i \in I(\hat{x}, \hat{y}). \quad (15)$$

Note, that the condition MFCQ in $A2_G$ (γ) is a natural condition implying that locally near \bar{x} the index set $Y(x)$ is continuous.

Again, we consider for fixed x under assumptions $A2_G$ the parametric problem

$$Q(x) : \quad \max_y g(x, y) \quad \text{s.t.} \quad v_i(x, y) \leq 0, \quad i \in I. \quad (16)$$

Under MFCQ at a solution \bar{y} of $Q(x)$ the Kuhn-Tucker condition must hold:

$$D_y g(x, \bar{y}) - \sum_{i \in I(x, \bar{y})} \mu_i D_y v_i(x, \bar{y}) = 0 \quad \text{with} \quad \mu_i \geq 0.$$

Lemma 6 *Suppose the assumption $A2_G$ holds. Let $\varepsilon_0 > 0$ be as in Lemma 5. Then, there exists $\alpha > 0$ such that for any x , $\|x - \bar{x}\| \leq \varepsilon_0$, $y \in Y(x)$ and $\mu \geq 0$ satisfying*

$$D_y g(x, y) - \sum_{i \in I(x, y)} \mu_i D_y v_i(x, y) = 0$$

we have $\|\mu\| \leq \alpha$.

PROOF : Suppose on the contrary, that there are sequences $\{x_k\}$, $\{y_k\}$, $\{\mu^k\}$, $\|x_k - \bar{x}\| \leq \varepsilon_0$, $y_k \in Y(x_k)$ and $\mu^k \geq 0$ such that

$$D_y g(x_k, y_k) - \sum_{i \in I(x_k, y_k)} \mu_i^k D_y v_i(x_k, y_k) = 0$$

and $\|\mu^k\| \rightarrow \infty$ for $k \rightarrow \infty$. Then as in the proof of Lemma 2, for accumulation points \hat{x} , $\|\hat{x} - \bar{x}\| \leq \varepsilon_0$, $\hat{y} \in Y(\hat{x})$, $\hat{\mu}_i \geq 0$, $i \in I_0$ and $I_0 \subset I(\hat{x}, \hat{y})$ we find

$$-\sum_{i \in I_0} \hat{\mu}_i D_y v_i(\hat{x}, \hat{y}) = 0$$

with $\hat{\mu} \neq 0$ and by multiplying this relation with the MFCQ-vector $\hat{\eta}$ in (15) we obtain a contradiction. \square

For $I_l \subset I$ we also can define the subsets of $Y(x)$,

$$f_l(x) := \{y \in Y(x) \mid v_i(x, y) = 0, i \in I_l\}, \quad x \in \mathbb{R}^n, \quad l = 1, \dots, k,$$

which now are depending on x .

A3_G : Suppose that the discretization Y_d is chosen such that for all $l = 1, \dots, k$, and all d with the set $N_{\bar{x}}$ in A1_G we have

$$\max_{y \in f_l(x)} \min_{\hat{y} \in Y_d(x) \cap f_l(x)} \|\hat{y} - y\| \leq d \quad \text{for all } x \in N_{\bar{x}}.$$

Theorem 4 *Let A1_G, A2_G and A3_G hold. Then with the local solutions x_d of GSIP(Y_d) for some $L > 0$ we have*

$$\max_{y \in Y(x_d)} g(x_d, y) \leq L d^2.$$

PROOF : The proof is a straightforward modification of the proof of Theorem 2. \square

Concerning a result as in Lemma 3, the situation is more complicated for semi-infinite problems with variable index sets. To prove such a result we have to strengthen the assumptions. We will assume that the so-called *reduction approach* is possible near \bar{x} (see e.g. [9], [14]).

Note that by definition of the active index set $Y_0(\bar{x}) = \{\bar{y} \in Y(\bar{x}) \mid g(\bar{x}, \bar{y}) = 0\}$, for feasible \bar{x} , any point $\bar{y} \in Y_0(\bar{x})$ is a global solution of the finite optimization problem $Q(\bar{x})$ in (16).

We make the following assumptions.

A_{red} : Let the functions g, v_i be twice continuously differentiable on $\bar{U} \times C_0$, where \bar{U} is a neighborhood of \bar{x} . Suppose for any $\bar{y} \in Y_0(\bar{x})$ (i.e. a solution \bar{y} of $Q(\bar{x})$):

(α) LICQ: $D_y v_i(\bar{x}, \bar{y}), i \in I(\bar{x}, \bar{y})$ are linearly independent.

(β) Kuhn-Tucker condition

$$D_y g(\bar{x}, \bar{y}) - \sum_{i \in I(\bar{x}, \bar{y})} \bar{\mu}_i D_y v_i(\bar{x}, \bar{y}).$$

with $\bar{\mu}_i > 0, i \in I(\bar{x}, \bar{y})$, (*strict complementary slackness*).

(δ) A certain second order sufficiency condition on the Hessian of the Lagrange function (see e.g. [14]).

It can be proven, that under these conditions A_{red} the set $Y_0(\bar{x})$ is finite, $Y_0(\bar{x}) = \{\bar{y}_1, \dots, \bar{y}_r\}$, and that there is a neighborhood \bar{U} of \bar{x} such that with local maximizer

functions $y_j(x) \in C^1(\bar{U})$, $y_j(\bar{x}) = \bar{y}_j$, $j = 1, \dots, r$, of $Q(x)$, locally on \bar{U} , GSIP is equivalent with the smooth finite problem (*reduced problem*):

$$\text{GSIP}_{\text{loc}}(\bar{x}) : \min f(x) \quad \text{s.t.} \quad g_j(x) := g(x, y_j(x)) \geq 0, \quad j = 1, \dots, r.$$

For the functions g_j the relation

$$Dg_j(\bar{x}) = D_x g(\bar{x}, \bar{y}_j) - \sum_{i \in I(\bar{x}, \bar{y}_j)} \bar{\mu}_i^j D_x v_i(\bar{x}, \bar{y}_j).$$

holds with the Lagrange-functions corresponding to the maximizer $\bar{y}_j \in Y_0(\bar{x})$.

The MFCQ has now to be modified: We say that MFCQ_G is valid for \bar{x} if there exists a vector $\xi \in \mathbb{R}^m$ such that

$$Dg_j(\bar{x}) \xi \leq -\kappa < 0, \quad \text{for all } j = 1, \dots, r. \quad (17)$$

Lemma 7 Suppose A_{red} , MFCQ_G is valid at the local minimizer \bar{x} of GSIP and with some $\gamma > 0$, $c_1 > 0$ we have

$$\max_{y \in Y(x_d)} g(x_d, y) \leq c_1 d^\gamma, \quad \text{for sufficiently small } d.$$

Then there exists $\rho > 0$ such that with the MFCQ-vector ξ in (17), for small d , the points $\hat{x}_d := x_d + \rho d^\gamma \xi$ are feasible for GSIP, i.e.

$$g(\hat{x}_d, y) \leq 0, \quad y \in Y(\hat{x}_d).$$

PROOF : By the reduction assumption A_{red} for x near \bar{x} for any $\sigma \in \mathbb{R}$ we have

$$g(x, y) \leq \sigma, \quad y \in Y(x) \quad \text{is equivalent with} \quad g_j(x) \leq \sigma, \quad j = 1, \dots, r.$$

Using the continuity of Dg_j , by MFCQ_G there exists $\varepsilon > 0$ such that

$$Dg_j(x) \xi \leq -\frac{\kappa}{2} \quad \text{for } j = 1, \dots, p, \quad x, \quad \|x - \bar{x}\| < \varepsilon.$$

Now we find for all $j = 1, \dots, r$,

$$\begin{aligned} g_j(\hat{x}_d) &= g_j(x_d) + \rho d^\gamma Dg_j(x_d) \xi + o(\rho d^\gamma) \\ &\leq c_1 d^\gamma - \rho \frac{\kappa}{2} d^\gamma + o(\rho d^\gamma) \\ &= d^\gamma (c_1 - \rho \frac{\kappa}{2}) + o(\rho d^\gamma) \leq 0 \end{aligned} \quad (18)$$

if we choose ρ , $\rho \frac{\kappa}{2} > c_1$ and d small enough. This shows that \hat{x}_d is feasible. \square

The generalization of Corollary 1 and Theorem 3 is again straightforward. We only give

Theorem 5 Suppose $A1_G$, A_{red} , $MFCQ_G$ hold and f is Lipschitz continuous at \bar{x} . Then if \bar{x} is a strict local minimizer of order $p = 1$ or $p = 2$ we have :

(a) There is some $\sigma_2 > 0$ such that for all d

$$\|x_d - \bar{x}\| \leq \sigma_1 d^{1/p}.$$

(b) If in addition $A2_G$ and $A3_G$ is satisfied for Y, Y_d , then with some $\sigma_2 > 0$ the inequality

$$\|x_d - \bar{x}\| \leq \sigma_2 d^{2/p}$$

is valid.

References

- [1] Cheney E.W., *Introduction to approximation theory*, McGraw-Hill, New York, (1966).
- [2] Dunham C.B., *Efficiency of Chebyshev approximation on finite subsets*, J. Assoc. Comp. Mach. 21, No. 2, 311-313, (1974).
- [3] Dunham C.B., Williams J., *Rate of convergence of discretization in Chebyshev approximation*, Math. of Computation 37, No. 155, 135-139, (1981).
- [4] Görner S., Reemtsen R., *Numerical methods for semi-infinite programming: A survey*, in 'Semi-infinite Programming', Reemtsen/Rückmann (eds.), Kluwer, Boston, p. 195 -262 (1998).
- [5] Hettich R., Zencke P., *Numerische Methoden der Approximation und semi-infiniten Optimierung*, Teubner, Stuttgart (1982).
- [6] Hettich R., *An implementation of a discretization method for semi-infinite programming*, Math. Programming 34 (1986), 354-361.
- [7] Hettich R., Gramlich G., *A note on an implementation of a method for quadratic semi-infinite programming*, Math. Programming 46 (1990), 249-254.
- [8] Hettich R., Kortanek K., *Semi-infinite programming: Theory, methods and applications*, SIAM Review, vol 35, No.3, 380-429, (1993).
- [9] Hettich R., Still G., *Second order optimality conditions for generalized semi-infinite programming problems*, Optimization Vol. 34, pp. 195-211, (1995).
- [10] Opfer G., *Solving complex approximation problems by semi-infinite optimization techniques: A study on convergence*, Num. Math. 39, 411-420, (1982).

- [11] Polak E., *Optimization, Algorithms and consistent Approximations*, Springer, Berlin (1997).
- [12] Rückmann J.-J., Stein O., *On linear and linearized generalized semi-infinite problems*, Annals of Operations Research, to appear.
- [13] Still G. *Vector-valued approximation problems; Numerical and genericity aspects*, Proceedings of the 3rd International Conference on Approximation and Optimization (Puebla, 1995), (16pp. electronic), Benemerita Univ. Auton. Puebla, Puebla, (1997).
- [14] Still G., *Generalized semi-infinite programming: Theory and methods*, European Journal of Operational Research 119, 301-313, (1999).
- [15] Still G., *Generalized semi-infinite programming: Numerical aspects*, Optimization, to appear